



Regional and intercontinental pollution signatures on 1 modeled and measured PAN at northern mid-latitude 2 mountain sites 3

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- 46 Abstract. Peroxy acetyl nitrate (PAN) is the most important reservoir species for nitrogen oxides (NO_x) in
- 47 the remote troposphere Upon decomposition in remote regions, PAN promotes efficient ozone production.





48 We evaluate monthly mean PAN abundances from global chemical transport model simulations (HTAP1) 49 for 2001 with measurements from five northern mid-latitude mountain sites (four European and one North 50 American). The multi-model mean generally captures the observed monthly mean PAN but individual 51 models simulate a factor of ~4-8 range in monthly abundances. We quantify PAN source-receptor 52 relationships at the measurement sites with sensitivity simulations that decrease regional anthropogenic 53 emissions of PAN (and ozone) precursors by 20% from North America (NA), Europe (EU), and East Asia 54 (EA). The HTAP1 models attribute more of the observed PAN at Jungfraujoch (Switzerland) to emissions 55 in NA and EA, and less to EU, than a prior trajectory-based estimate. The trajectory-based and modeling 56 approaches agree that EU emissions play a role in the observed springtime PAN maximum at Jungfraujoch. 57 The signal from anthropogenic emissions on PAN is strongest at Jungfraujoch and Mount Bachelor 58 (Oregon, U.S.A.) during April. In this month, PAN source-receptor relationships correlate both with model 59 differences in regional anthropogenic volatile organic compound (AVOC) emissions and with ozone 60 source-receptor relationships. PAN observations at mountaintop sites can thus provide key information for 61 evaluating models, including links between PAN and ozone production and source-receptor relationships. 62 Establishing routine, long-term, mountaintop measurements is essential given the large observed 63 interannual variability in PAN.

64 **1 Introduction**

65 Ozone (O_3) distributions over northern mid-latitude regions reflect a combination of O_3 produced from sources within a given region and O₃ transported from outside that region. Intercontinental influence on 66 67 surface O_3 levels occurs via at least two pathways: (1) O_3 can be produced within a polluted continental 68 boundary layer, ventilated to the free troposphere and efficiently transported to other continents; and $(2) O_3$ 69 can also be produced in transit from the export and subsequent chemical evolution of precursors such as 70 peroxyacetyl nitrate (PAN) (Jaeglé et al., 2003; Kotchenruther et al., 2001a; Hudman et al., 2004). PAN 71 serves as a reservoir species for nitrogen oxides (NO_x), facilitating long-range transport of this chemical family. To better understand differences in the quantification of O_3 transport, the Task Force on 72 73 Hemispheric Transport of Air Pollution (HTAP) organized a global model intercomparison exercise, 74 referred to here as HTAP1. A factor of two difference was identified in the HTAP1 model estimates of 75 surface O₃ response to changes in anthropogenic precursor emissions from continental-scale, northern mid-76 latitude source regions (HTAP 2007; HTAP 2010). These precursor emissions include both NO_x and non-77 methane volatile organic compounds (VOC) that produce PAN along with O_3 . Here, we take a first step 78 towards determining the extent to which source-receptor relationships for PAN and O3 are related. 79 Specifically, we extend the prior HTAP1 analysis to evaluate lower tropospheric PAN distributions and to 80 quantify the impact of regional versus intercontinental anthropogenic emissions on PAN abundances. 81 Improved understanding of the processes controlling PAN abundances will help quantify the impact of 82 intercontinental transport on surface O3.





84 PAN is formed from the reversible reaction between acetyl peroxy radicals (CH₃COO₂) and nitrogen 85 dioxide (NO₂). Once formed, PAN can be lofted to the free troposphere, where it is more stable at colder 86 (lower) temperatures and can be efficiently transported throughout the hemisphere (Singh, 1987; Singh and 87 Hanst, 1981). The lifetime of PAN against thermal decomposition is about 1 hour at 20°C, and it approximately doubles for every 4°C decrease in temperature. In the mid-troposphere during spring, when 88 89 intercontinental air mass flow is strongest, the lifetime of PAN is longer than a month. When PAN decomposes in low-NO_x regions of the atmosphere, the NO_x released can produce O₃ up to eight times 90 91 more efficiently than in polluted (high-NO_x) regions (Liang et al., 1998; Liu et al., 1987). Several studies emphasize the potential for PAN to increase the hemispheric NO_x burden and thus global O_3 abundances 92 93 (Moxim et al., 1996; Wang and Jacob, 1998), as O₃ formation is NO_x-limited in most of the free 94 troposphere (Chameides et al., 1992). Owing to the sensitivity of PAN thermal decomposition to 95 temperature, PAN also plays a role in altering the distribution of reactive nitrogen and hence O₃ under a warming climate (e.g. Doherty et al., 2013). Efficient O₃ production in the lower troposphere following 96 97 subsidence of PAN-containing air masses has been observed during several aircraft field campaigns in the Eastern Pacific and at mountain top sites in the Western U.S. and North Atlantic (Fischer et al., 2010; 98 99 Heald et al., 2003; Hudman et al., 2004; Kotchenruther et al., 2001a,b; Val Martin et al., 2008a; Zhang et al., 2008). 100

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The role of O_3 formation resulting from PAN chemistry by the two main pathways described above – via 102 103 PAN chemistry versus direct transport of O_3 formed in a remote boundary layer – in contributing to the 104 estimates of intercontinental source-receptor relationships for surface O₃ reported by HTAP (2010) is not 105 well quantified. Jaegle et al. (2003) provided a budget for the Pacific Northwest, suggesting that 28% of the 106 O₃ is associated with PAN-to-NO_x conversion in the Pacific free troposphere between 0-6 km. One regional 107 modeling study found that PAN export and its subsequent decomposition, followed by NOx-induced O3 108 formation, contributed 20% of the spatially averaged response of surface O3 levels over East Asia to 109 European anthropogenic emission changes to (Lin et al., 2010). The same study found that this process may 110 contribute up to 50% of the O₃ response to European emission changes at mountain sites. A global 111 modeling study found that PAN produced from East Asian emissions and exported to the free troposphere 112 contributes 35% and 25% in spring and summer, respectively, to the free tropospheric O₃ abundance over 113 western North America (Jiang et al., 2016).

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We hypothesize that PAN may serve as a sensitive diagnostic of model uncertainties in O_3 production chemistry and transport (Emmerson and Evans, 2009; Kuhn et al., 1998), particularly given that model errors in O_3 production may be buffered by compensating errors in O_3 losses. Prior analysis of measurements and global model simulations suggests that PAN abundances at high altitude sites may be more sensitive than O_3 itself to changes in precursor emissions (Fiore et al., 2011; Fischer et al., 2011; Jaffe et al., 2007). We interpret this stronger sensitivity of PAN to changes in precursor emissions, as compared





121 to O₃, in terms of the relative independence of PAN loss pathways to changes in emissions. While PAN 122 production loosely mirrors much of the complex O₃ production chemistry, its dominant loss pathways are 123 much simpler than those for O₃. These PAN loss pathways are less likely to respond to changes in 124 emissions than those for O₃. PAN loss pathways include thermal decomposition (which dominates below approximately 7 km); photolysis in the upper troposphere; and dry deposition within the boundary layer 125 (Kirchner et al., 1999; Roberts, 2007; Turnipseed et al., 2006). All of the HTAP1 models include PAN 126 127 formation, but the chemical mechanisms and kinetic rate coefficients differ, with likely implications for long-range transport (Emmerson and Evans, 2009). More generally, PAN is a useful tracer of 128 129 photochemistry because of its low background and the absence of direct emissions.

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131 For our analysis, we construct climatological monthly mean PAN abundances by compiling multi-vear 132 PAN measurements at northern mid-latitude mountain sites and use them to evaluate monthly mean abundances simulated with the models that participated in the HTAP1 project. We view this analysis as a 133 first step towards determining the extent to which model differences in PAN can provide insights into 134 135 model representation of processes that also influence intercontinental O₃ transport. While aircraft 136 observations have advanced the understanding of the chemistry and dynamics of individual plumes (e.g., Alvarado et al., 2010), their limited temporal coverage is not well suited for comparison with the HTAP1 137 monthly mean PAN mixing ratios. Similarly, the HTAP1 monthly mean mixing ratios are not well suited to 138 139 evaluation with tropospheric PAN column abundances retrieved from instruments aboard satellite platforms 140 (e.g., Payne et al., 2014) as these require application of retrieval-specific averaging kernels and a priori 141 profiles. These other observational platforms are likely to provide additional constraints on inter-model discrepancies in coordinated multi-model efforts that archive three-dimensional fields at daily or higher 142 temporal frequency at the locations of aircraft flights and satellite overpasses. 143

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We begin by describing the HTAP1 model simulations and mountaintop measurements (Section 2) before 145 146 quantifying the multi-model range of lower tropospheric PAN distributions, and evaluating this range with available measurements at northern mid-latitude mountain sites (Section 3). We then focus on the 147 148 Jungfraujoch site in Europe to capitalize on a prior attribution of measured PAN mixing ratios to European, 149 North American, and East Asian emissions using trajectory-based methods (Pandey Deolal et al, 2013), and 150 compare these estimates with those derived from model sensitivity simulations (Section 4). We select one 151 European and one North American mountain site to probe relationships between simulated PAN and 152 regional emissions or transport in the individual models, with the goal of identifying some of the dominant 153 processes controlling hemispheric transport of PAN to these locations (Section 5) and examine the potential 154 relevance of conclusions regarding PAN origin to O₃ (Section 6). Finally, we summarize our conclusions 155 and offer a few recommendations for future work (Section 7).





156 2. Approach

157	In this section, we describe the HTAP1 model simulations (Section 2.1), the multi-year measurements of
158	PAN at five northern mid-latitude mountaintop sites, and our strategy to sample the models at these sites
159	(Section 2.2). We then describe the methodology used by Pandey Deolal et al. (2013) to derive trajectory-
160	based attributions of PAN measured at Jungfraujoch to East Asian, European, and North American source
161	regions, and discuss some limitations of this approach (Section 2.3). We use these trajectory-based source
162	attributions to evaluate the responses of PAN at Jungfraujoch to regional emission perturbations in the
163	HTAP1 model simulations.

164 2.1 HTAP1 model simulations

We use monthly mean PAN mixing ratios simulated by fourteen global chemistry transport models for the 2001 meteorological year (Table 1); the temporal resolution available for most three-dimensional chemical fields in the HTAP1 model archive is limited to monthly. While simulations from Phase 2 of HTAP (HTAP2) are available, they focus on meteorological years of 2008-2010. Given our interest in comparing with the trajectory-based attribution of observations in 1997-1998 (Section 4) and the substantial changes in NO_x emissions (a precursor to PAN) between the late 1990s and late 2010s (e.g., Hilboll et al., 2013), we restrict our analysis to the HTAP1 year 2001 simulations.

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We consider four HTAP1 Source-Receptor (SR) simulations (Table 2): a base case (denoted SR1, SR 173 simulation 1, for consistency with prior publications) and three perturbation simulations in which 174 175 anthropogenic O_3 precursor emissions (NO_x, VOC, carbon monoxide and aerosols) for the year 2001 are 176 reduced simultaneously by 20% within East Asia (SR6EA), Europe and northern Africa (SR6EU), and 177 North America (SR6NA). These three regions, depicted as white boxes in Figure 1, have been shown to 178 influence lower tropospheric O₃ (and PAN) at northern mid-latitudes more than the fourth HTAP1 region 179 (SR6SA; Jonson et al., 2010). We attribute PAN, sampled at mountaintop sites (see Section 2.2), separately 180 to emissions within these three source regions by differencing the perturbation and base simulations 181 (SR6XX-SR1, where XX refers to the region in which emissions of PAN precursors were decreased by 20%). We refer to this difference as the source-receptor relationship for PAN. 182

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Of the models in Table 1, eleven used prescribed meteorological fields to drive tracer transport for the year 2001. Two models are chemistry-transport models coupled directly to a general circulation model forced by observed sea surface temperatures (STOC-HadAM3 and STOCHEM) and one model incorporates chemistry directly into a general circulation model (UM-CAM). We include results from these three models here as their climatological nature matches that of our evaluation, which compiles PAN measurements across several years. The individual model specifications and emissions are described in detail in Fiore et al. (2009), with a summary of specifications and simulations relevant for our study provided in Tables 1





and 2. In HTAP Phase I, each model used its own emissions inventories (listed in Table A1 of Fiore et al., 2009); Fiore et al. (2009) give emission totals within each HTAP Phase 1 source region for all (their Table A2) and anthropogenic (their Table A3) emissions of NO_x, NMVOC, and CO. The relative inter-model spread in regional anthropogenic emissions is smallest for NO_x emissions in EU and NA (<10%) and largest for VOC from EU (58%) (Fiore et al. 2009). We conduct one additional simulation in the MOZARTGFDL-v2 model, in which we increase North American isoprene emissions by 20% (ISOPNA; described in Fiore et al., 2011), to demonstrate the sensitivity of PAN formation to isoprene emissions.

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199 To separate the role of differences in model transport on simulated PAN from the combined impacts of differences in model emissions and chemistry, we analyze an additional set of idealized tracer simulations 200 201 available from eleven models (see column COfromXX in Table 1, where XX is the source region). In these simulations, a set of tagged carbon monoxide-like tracers are emitted, each from a single HTAP1 source 202 203 region with a 50-day lifetime, and with identical emissions across models. Biomass burning emissions for the CO tracers are from GFED (van der Werf et al., 2006; 2010) and other emissions are from the RETRO 204 project (Schultz et al., 2007; 2008). We refer to these tracers as "COfromEA", "COfromNA", and 205 "COfromEU", which denote the tracers emitted from EA, NA, EU, respectively (Table 2; see also Doherty 206 207 et al., 2013; Shindell et al., 2008).

208 2.2 Multi-year PAN measurements at mountaintop sites and model sampling

209 To evaluate the HTAP1 models, we compiled monthly climatologies of lower tropospheric PAN 210 measurements from northern mid-latitude mountain observatories (Table 3). Given the large interannual 211 variability in PAN abundances, we restricted our analysis to sites with at least two years of observations in 212 the same month, from which we estimate a monthly climatology for PAN to compare to the models. PAN 213 observations from Mount Bachelor (U.S.A.), Jungfraujoch (Switzerland), and Zugspitze 214 (Schneefernerhaus), Hohenpeissenberg, and Schauinsland (all in Germany) meet these criteria (Table 3). 215 Site locations are shown as white circles in Figure 1. Taken together, these mountaintop measurements 216 span 15 years, from 1995 to 2010, although only one site (Schauinsland) overlaps with the year 2001 217 chosen for the HTAP1 simulations.

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PAN was measured at all five mountain sites using gas chromatography with electron capture detection (ECD). A custom system using a Shimazu Mini-2 ECD was employed at Mount Bachelor (Fischer et al., 2010) and the commercially available Meteorologie Consult (GmbH) system was used at the European sites (Zellweger et al., 2000). Calibrations were performed using a photochemical source where PAN is generated from the photolysis of excess acetone and NO in air (Warneck and Zerbach, 1992; Volz-Thomas et al., 2002). Reported detection limits are ~20 ppt for PAN measurements at Mount Bachelor, and ~50 ppt for the European sites, with total uncertainties of <10% (Fischer et al., 2010; Zellweger et al., 2003).</p>





227 We use monthly mean values at each of the measurement sites in order to match the temporal resolution of 228 PAN archived from the models. We include all available data at each site to calculate monthly means, and 229 do not filter the data for upslope winds or any other criteria. When measurements are below detection limit, 230 half the detection limit is used to calculate mean values. This assumption should not affect our conclusions as PAN values below detection limit are generally observed in nocturnal boundary layers were PAN is 231 232 depleted by deposition whereas these high elevation sites generally sample free tropospheric air at night 233 (e.g., Weiss-Penzias et al., 2004). At Mount Bachelor, the cleanest of the 5 mountaintop sites used in 234 Figure 2, Fischer et al. (2010) have shown that the range of PAN measurements during springtime (our 235 focal season) is an order of magnitude larger than the amplitude of the diurnal cycle in PAN, suggesting 236 that PAN mixing ratios are not primarily controlled by diurnal wind patterns.

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238 For comparison with the observations, we sample each model on its native grid (Table 1) at the horizontal 239 grid cell containing the latitude and longitude of each mountain site. Orography at these mountain sites is 240 poorly resolved at the relatively coarse HTAP1 model horizontal resolutions, and consequently the model 241 vertical grid differs from the real world. This mismatch requires us to apply some approximations for 242 sampling in the vertical. We convert the station altitude to an approximate pressure level by assuming a 243 mean tropospheric temperature of 260 K, which gives an atmospheric scale height of 7.6 km. We then 244 sample the models by linearly interpolating between the pressures of the two model grid cells that vertically 245 bound this pressure level, using the corresponding monthly pressure fields from each model. While different sampling strategies may alter the exact value of simulated PAN and its comparison to 246 247 observations, our primary interest is in the inter-model differences. Although the Zugspitze and 248 Hohenpeissenberg sites fall within the same horizontal grid cell due to the coarse horizontal resolution of 249 the HTAP1 models, the station altitudes differ, so we consider the two sites separately.

250 **2.3** Trajectory-based attribution of PAN measurements at Jungfraujoch to emission region

251 Trajectories were calculated from 6 hourly wind fields from ECMWF (European Centre for Medium-Range 252 Weather Forecasts) ERA Interim reanalysis data. The trajectory-based source attribution to a given region is determined by averaging PAN mixing ratios every six hours on days when 20-day back-trajectories can 253 254 be unambiguously assigned to the continental source region based on recent boundary layer contact. The values for PAN attributed to a given source region represent the mean mixing ratios on days with 255 256 trajectories attributed to the boundary layer over that source region, weighted by the frequency of 257 trajectories originating in that source region as described by Pandey Deolal et al. (2013). The observed fractional attribution is estimated as the frequency-weighted values divided by the observed monthly 258 259 means.

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A limitation of this trajectory-based source attribution is that the lifetime of PAN often exceeds the time scale over which the back trajectories can be used to trace PAN observed at Jungfraujoch back to its





263 original continental boundary layer source. Pandey Deolal et al. (2013; see their Figure 1) label these cases 264 as "free troposphere" and find that 13-32% (range depends on season and year) of all trajectories fall in this 265 category. Uncertainties in trajectory calculations normal to the direction of flow are typically about 20% of the horizontal distance traveled and thus increase with time (Pandey Deolal 2013; Stohl, 1998). As 266 trajectory uncertainties vary from case to case, an ensemble of backward trajectories is calculated at 267 268 Jungfraujoch to reduce error. Given the dependence of uncertainty on trajectory length, constraints from the 269 trajectory-based approach are likely to be stronger for the regional European emission influence than for 270 the more distant source regions of North America and East Asia.

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272 The trajectory-based approach assumes that all PAN in an air mass that originates over a continent is 273 produced from PAN precursor emissions (NO_x and VOC) over that continent and may thus overestimate the 274 influence from that particular source region if there are non-trivial enhancements to PAN in the air mass 275 produced by emissions in other regions. On the other hand, the "free troposphere" air masses must contain some influence from global PAN precursor emissions, which would tend to cause the trajectory-based 276 277 approach to underestimate the influence from any source region. In contrast, the HTAP1 model 278 perturbation simulations directly diagnose the impact of precursor emissions within a given source region to the PAN simulated at the Jungfraujoch site. We interpret the trajectory-based attributions as likely 279 offering a lower limit for the intercontinental influence from NA and EA at Jungfraujoch and an upper limit 280 281 for PAN originating from the EU source region. Despite the limitations outlined here, this trajectory-based 282 approach offers one of the only available independent methods for evaluating source-receptor relationships 283 diagnosed with the global HTAP1 models. Evaluating emission-response relationships is critical to build 284 confidence in using these models to project responses to future emission changes.

285 3. Modeled and measured lower tropospheric PAN at northern mid-latitudes

At northern mid-latitudes, intercontinental pollutant transport maximizes in spring (e.g., HTAP 2010). 286 287 Earlier work attributed the observed springtime maximum in tropospheric PAN to the build-up of longerlived precursors during the northern hemisphere winter and subsequent PAN production during spring as 288 289 photochemical activity rises; at the same time, colder temperatures during spring slow the PAN decomposition rate relative to the warmer summer months (Bottenheim et al., 1994; Penkett and Brice, 290 1986; Schmitt and Volz-Thomas, 2004). At Jungfraujoch, the highest springtime PAN levels have been 291 292 shown to occur in air masses originating in the European planetary boundary layer (Pandey Deolal et al., 293 2013, 2014). This maximum in EU influence reflects strong photochemical activity in spring, enhanced 294 thermally induced vertical transport from the European boundary layer, and a longer PAN lifetime against 295 thermal decomposition relative to the warmer summer season. Smaller springtime maxima occur in free 296 tropospheric air masses, implying a larger role for regional pollution in contributing to the observed 297 springtime maximum in PAN than previously recognized (Pandey Deolal et al., 2013, 2014).





299 We examine simulated PAN distributions during April and July in the northern hemisphere to illustrate 300 characteristic differences in the spring versus summer distributions as represented by the model ensemble 301 mean PAN mixing ratios at 650 hPa (~3 km), the level sampled by the three highest altitude sites in Table 302 2. Higher mixing ratios and a stronger latitudinal gradient are evident in April as compared to July (Figure 1 left column). Compared to April, the lower PAN mixing ratios downwind of source regions in July reflect 303 304 the shorter lifetime of PAN at warmer temperatures. In both seasons, enhancements occur over the major 305 source regions of the eastern United States and Mexico, Europe, and Asia. The southward extension of 306 higher PAN mixing ratios over North America in July as compared to April reflects (1) the impact of 307 biogenic PAN precursors, such as emission of isoprene and its subsequent oxidation to products including 308 methyl glyoxal and acetaldehyde (Fischer et al., 2014; Roberts et al., 2006; Wolfe et al., 2009), and (2) 309 active convection and the associated lightning NO_x in the warmer summer months.

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The multi-model relative standard deviations (the coefficient of variation: the standard deviation over the 311 14 models divided by the model ensemble mean) are also shown in Figure 1 (right column; note we set a 312 313 100 ppt PAN threshold level to emphasize variability across the models in locations with higher PAN 314 mixing ratios). The multi-model spread in lower tropospheric PAN (at 650 hPa) in April mainly follows a 315 latitudinal gradient, with less inter-model deviation at high-latitudes where PAN is most abundant and more inter-model deviation at lower latitudes where PAN mixing ratios are lower. The higher inter-model spread 316 in remote regions reflects not only differences in model representations of PAN formation, but also 317 318 differences in transport and subsequent thermal decomposition. In April, the inter-model relative standard 319 deviations over Europe exceed 50%, suggesting that the European mountain sites may be particularly 320 useful to test the models. In summer, variability across the models is largest in the North Pacific off the Alaskan Aleutian islands, off the west coast of Mexico, and over the Atlantic centered at about 30°W and 321 30°N, near the Pico Mountain, Azores site (e.g., Honrath et al., 2004; Val Martin et al., 2008ab). Future 322 323 measurements in regions where model variability is large would provide constraints on PAN distributions 324 in the models, and on the relative influence from different source regions on PAN, as discussed below.

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Observed PAN mixing ratios at the northern mid-latitude mountain sites tend to be highest in spring 326 327 (Figure 2), consistent with remote measurements at lower elevations in the northern hemisphere 328 (Bottenheim et al., 1994; Penkett and Brice, 1986; Singh and Salas, 1989). Substantial year-to-year 329 variability (black vertical bars in Figure 2) is evident from the measurements (Zellweger et al., 2003; 330 Fischer et al., 2011; Pandey Deolal et al. 2013, 2014). Capturing this interannual variability would be a useful test for future multi-year modeling efforts. On average, the measurements at all sites (black circles) 331 332 indicate that the spring peak occurs in April. An exception occurs at Schauinsland in 2001 (blue triangles), 333 when the peak is delayed until May, and a secondary peak occurs in August. Consistent with the 334 measurements, the 2001 multi-model mean is highest in April at Mount Bachelor and Zugspitze, but occurs 335 in May at Schauinsland, although in early spring the multi-model mean abundance falls closer to the multi-





336 year average of the measurements than the low values measured in 2001. Jungfraujoch lacks measurements 337 in 2001 but the measured inter-annual variability is largest in May. The large observed variability at 338 Jungfraujoch indicates that there are individual years where PAN at Jungfraujoch peaks in May. The models suggest that a May peak occurred during 2001 at Jungfraujoch, similar to the observations and 339 multi-model mean at Schauinsland. At Hohenpeissenberg, the measurements peak in April, but the models 340 341 indicate a broad spring-summer maximum with higher PAN abundances than measured as well as a 342 September peak. This low elevation site may be more susceptible to the influence of boundary layer PAN formation from European emissions (e.g., Gilge et al., 2010; Kaiser et al., 2007; Naja et al., 2003; Pätz et 343 344 al., 2000), and we interpret the model overestimate at Hohenpeissenberg as evidence that the models cannot properly resolve the vertical gradients in PAN. This shortcoming could in part reflect inadequate dry 345 346 deposition (e.g., Turnipseed et al., 2006), and thus an overestimate of European boundary layer influence on the magnitude and seasonality of PAN abundances. 347

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349 Individual model simulations indicate a wide spread across the models (gray shading in Figure 2). 350 Systematic biases versus the observations are evident in some models. For example, the CAM-Chem model 351 simulates PAN abundances that are usually higher than observed mixing ratios (brown line in Figure 2), and than most of the other models, whereas PAN abundances simulated by GISS-modelE are always low 352 353 during summer (green line in Figure 2). Monthly mean PAN abundances in the GMI-v0f (cyan) and MOZART2-GFDL (red) models, however, do not consistently follow the same ranking from site to site. 354 355 Overall, the multi-model mean (dashed line) generally falls close to the monthly mean observed mixing 356 ratios with a tendency towards a positive bias at Mount Bachelor in all seasons (within ~50 ppt) and at 357 Hohenpeissenberg in the warm season.

358 4. Evaluation of regional and intercontinental influences on PAN at Jungfraujoch

359 We apply the trajectory-based estimates of the contributions from emissions within EU, NA, and EA to PAN measured at Jungfraujoch (Pandey Deolal et al., 2013; described in Section 2.3) to evaluate the PAN 360 361 source-receptor relationships simulated by the HTAP1 models. For this comparison, we multiply the PAN source-receptor relationships (SR1-SR6XX; Section 2.1 and Table 2) by a factor of 5 to scale from the 20% 362 363 emission perturbations to 100%. Our linear scaling of the PAN response from 20% to 100% emission 364 perturbations may incur errors due to non-linear chemistry. To assess the magnitude of this error, we analyze an additional simulation with the FRSGCUCI model that sets European anthropogenic emissions of 365 NO_x, CO and VOC to zero (a 100% perturbation). We find that assuming linearity in the PAN response to 366 European emission changes (20%*5 vs. 100%) incurs an error of ~10% over the EU emission region and 367 <3% over other regions. Earlier work assessing linearity in O3 responses also indicates stronger non-368 369 linearity to regional versus intercontinental sources (Fiore et al., 2009; Wu et al., 2009; Wild et al., 2012). 370 Although PAN and O_3 share the same precursor emissions, O_3 responds more strongly to changes in NO_x 371 emissions outside of high-NO_x regions, whereas PAN is sensitive to changes in both NO_x and VOC





emissions, and in many regions PAN responds more strongly to changes in VOC emissions (Fischer et al., 2014; see their Figure 4). The sensitivity of PAN to changes in VOC emissions can be amplified by a chemical feedback in remote regions where PAN is the main NO_x source: in these regions, if PAN decreases then O_3 decreases, which lowers the $[NO_2]/[NO]$ ratio that controls PAN formation, thereby reducing the effective PAN lifetime (Fischer et al., 2014).

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378 Figure 3 shows the observation-derived source attributions with those estimated by the model ensemble, for both absolute (left column) and relative (right column) influences of precursor emissions from EU, NA, and 379 380 EA on PAN. The relative (fractional) contributions are obtained by dividing the absolute source-receptor 381 relationships by the total PAN mixing ratio in the SR1 simulation. Data in months with multiple years of 382 observations (April and May) show that the trajectory-based estimates for a given emission region can vary from year to year by up to a factor of two for the PAN abundance attributed to that region, and by over 20% 383 384 for the relative attribution. The HTAP1 model estimates span an even larger range than the trajectory-based 385 estimates from two different years.

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Of all the back-trajectories from Jungfraujoch, up to 40% originated in EU, 10-20% originated in North 387 America and fewer than 4% originated in EA (Pandey Deolal et al., 2013), consistent with the relative 388 ranking of multi-model mean source-receptor relationships for PAN from these two regions in Figure 3. 389 390 The multi-model mean is generally lower than the trajectory-based attribution to EU in late spring and 391 summer, possibly resulting from an inadequate model representation of local valley-mountain (upslope) 392 flows, which typically peak during these seasons (Pandey Deolal et al., 2013). For the intercontinental 393 source regions, all of the models exceed the estimates from the trajectory-based approach outside of the 394 warm season, with the exception of PAN from EA during October. Overall, the absolute trajectory-based 395 attributions of PAN at Jungfraujoch to intercontinental emissions fall close to the multi-model mean 396 estimate for the NA region from May through September and for the EA region in October (Figure 3). 397 Simulated fractional contributions of PAN produced from NA anthropogenic emissions are higher than the trajectory-based estimates (with the exception of September) as expected given the inability of the 398 399 trajectory-based approach to attribute all measurements to a source region. Nevertheless, if the trajectory-400 based attribution in Figure 3 provides a lower limit for the EA and NA source regions and an upper limit 401 for the EU source region (Section 2.3), then these two approaches (trajectory-based versus perturbation simulations in the HTAP1 models) to determining the influences of regional versus intercontinental sources 402 403 on PAN measured at Jungfraujoch are consistent.

404

The higher PAN abundances in the HTAP1 models during the cold season reflect the long lifetime of PAN at cold temperatures, which far exceeds the 20-day trajectory time scale. When the PAN lifetime exceeds 20 days, the trajectory-based approach cannot link the PAN measured at Jungfraujoch to a specific continental boundary layer. We attribute the apparent mismatch in the seasonal cycle in the NA influence





409 on PAN at Jungfraujoch (Figure 3) to this shortcoming of the trajectory-based approach. Given the 410 substantial year-to-year variability in the monthly mean measurements, stronger constraints could be placed 411 on inter-model discrepancies if the model and trajectory-based attribution approaches used the same 412 meteorological year.

413

414 The absolute range of the PAN source-receptor relationships in the HTAP1 models is wide for all three 415 source regions (Figure 3, left). The inter-model spread in the fractional attribution is smaller for NA than 416 for the nearby EU region (Figure 3, right). The models rank differently in their estimates for absolute 417 versus relative attribution of PAN to anthropogenic emissions from the three regions. For example, the 418 typically high (brown curves in Figure 3) and lower (green curves in Figure 3) models agree more closely 419 on the relative attribution despite the large discrepancies in their simulated total PAN abundances (Figure 420 2) and in the absolute amount of PAN attributed to EU and NA sources. We conclude that constraints on 421 both absolute and relative ranges of source-receptor relationships are useful as some of the inter-model range in PAN source-receptor relationships reflects differences in the mean state simulations that cancel out 422 423 in the fractional estimates.

424

425 In the future, as climate warms and anthropogenic emission controls continue to be phased in, we expect an 426 increasing role for temperature-sensitive PAN precursors such as biogenic VOC (BVOC). As an example, 427 we quantify the relative role of BVOC versus AVOC in contributing to intercontinental PAN source-428 receptor relationships at Jungfraujoch in 2001. Specifically, we use an additional perturbation simulation 429 from a single HTAP1 model (MOZART2-GFDL; shown in red in Figures 2 and 3) in which isoprene 430 emissions were increased by 20% over NA, as might occur as climate warms, and compare the influence of 431 BVOC on PAN with the influences determined from the 20% decreases to anthropogenic emissions from the three mid-latitude HTAP1 source regions (i.e., ISOPNA-SR1 for comparison with the magnitude of 432 433 SR1-SR6XX; we multiply these differences by a factor of 5 to estimate a 100% contribution in Figure 4). 434 In this model, a 20% increase in NA isoprene emissions exerts an influence on PAN at Jungfraujoch roughly equivalent to a 20% reduction to NA anthropogenic NOx, CO and VOC emissions from June-435 436 November, or to a 20% decrease in EU anthropogenic emissions in June and August (Figure 4). The 437 intercontinental isoprene influence exceeds that from EA anthropogenic emissions throughout the warm 438 season. Fischer et al. (2014) find that contributions to PAN over Europe from BVOC emissions and from 439 lightning NO_x both peak in summer in the GEOS-Chem model (see their Figure 7). Other studies have also 440 noted strong impacts on PAN from lightning and biomass burning emissions (e.g., Fadnavis et al., 2014; 441 Fischer et al., 2011; Real et al., 2010; Zhu et al., 2015). We conclude that interpretation of both 442 summertime measurements and differences across models requires consideration of PAN precursors from 443 biogenic, lightning, and biomass burning sources. Springtime measurements of PAN, however, should 444 reflect more directly the influence of anthropogenic emissions.





445 5. Factors contributing to the inter-model range in PAN source-receptor relationships

446 In this section, we aim to advance our understanding of the factors contributing to model differences in the 447 PAN response to continental-scale emission changes. Our approach is to (1) examine the spread in sourcereceptor relationships for PAN at Mount Bachelor in North America and Jungfraujoch in Europe, and (2) 448 449 link these to model differences in emissions or transport. We first identify months where the anthropogenic 450 emission influence on PAN is strongest. The HTAP1 multi-model ensemble mean values indicate that 80-451 86% of the total multi-model mean PAN abundances at the two sites in January through April are produced 452 from anthropogenic emissions (NO_x, VOC, and CO) within the three northern mid-latitude source regions 453 (Figure 5). During boreal summer, this contribution drops to only 40-60% (Figure 5) due to the additional 454 influences from lightning NO_x, biogenic, and biomass burning emissions of PAN precursors (e.g., see NA 455 isoprene source in Figure 4). These warm season sources complicate our efforts to disentangle the role of 456 inter-model differences in anthropogenic emissions versus transport on simulated PAN source-receptor relationships. In April, however, intercontinental pollution influence is highest in the multi-model ensemble 457 458 (the minimum in the difference between the dashed and solid black lines in Figure 5). The inter-model 459 differences in PAN abundances thus contain the strongest signals tied to inter-model variations in anthropogenic precursor emissions in April. We focus our remaining analysis on April to probe more 460 461 deeply the inter-model differences in PAN responses at the two mountaintop sites to changes in 462 anthropogenic emissions.

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464 We assess the importance of model differences in regional anthropogenic NO_x and NMVOC inventories 465 and transport in contributing to the spread across the simulated source-receptor relationships (gray shading in Figure 3 at Jungfraujoch). At each site, we examine the correlation between simulated PAN attributed by 466 each model to anthropogenic emissions of VOC (AVOC; Figure 6) or NOx. We find that the inter-model 467 468 range in a given regional AVOC emissions explains as much as 64% of the variation in PAN attributed to EU emissions, and about 25% of the variance in PAN attributed to the NA region (Figure 6). The 469 relationships for the EA source-receptor relationships are not significant. In contrast to AVOC, we find 470 little relationship between the range in simulated PAN source-receptor relationships at the mountain sites 471 472 and the model spread in regional ANO_x emissions. Differences in model transport (e.g., Arnold et al., 2017; 473 Orbe et al., 2017) may also contribute to the inter-model differences in PAN source-receptor relationships. 474 Our analysis of the HTAP1 idealized CO tracers (Section 2.3) reveals little correlation between inter-model 475 differences in these idealized tracers and in PAN source-receptor relationships sampled at Mount Bachelor 476 and Jungfraujoch. Fischer et al. (2014) have previously shown that PAN abundances respond more strongly 477 to changes in emissions of VOC than of NOx. Our analysis supports that earlier finding and highlights a key 478 role for model differences in regional AVOC emissions in contributing to the inter-model range in PAN 479 source-receptor relationships.





481 To investigate further these inter-model differences in PAN source-receptor relationships, we consider next 482 the importance that one model ascribes to a given source region relative to another source region. We first 483 examine the relative importance of emissions within the source region where the measurement site is 484 located versus the upwind intercontinental source region on PAN, by comparing NA:EA at Mount Bachelor and EU:NA at Jungfraujoch. At Mount Bachelor, the HTAP1 multi-model mean source-receptor 485 486 relationships from NA, EA, and EU, are roughly equal in April (Figure 5). The differences across the 487 HTAP models in the relative importance of the NA:EA source regions on PAN (which range from about 488 0.5 to 2.5) correlate roughly equally with the ratio of the NA:EA transport tracers and with the ratio of the 489 NA:EA AVOC emissions (Spearman rank correlation coefficient (r) = 0.6 for both cases; Figure 7 left 490 column). We find no relationship with the ratio of the NA:EA ANO_x emissions. At Jungfraujoch, the 491 HTAP1 multi-model mean attributes much of the PAN to emissions from the EU and NA source regions 492 during April (Figure 5). The ratio of PAN attributed to EU versus NA at Jungfraujoch, however, varies 493 from approximately 0.5 to 2 across the individual HTAP1 models (Figure 7; right column). In contrast to 494 our findings at Mount Bachelor, this ratio at Jungfraujoch depends most strongly on the ratio of ANO_x 495 emissions in the EU to NA regions (r = 0.6), and more weakly on the ratio of EU:NA AVOC emissions (r = 496 0.5). The correlation is even weaker between the ratio of PAN source-receptor relationships for these two 497 regions with inter-model differences in transport as diagnosed with the CO tracers from EU versus NA (r = 0.4). 498

499

500 We next examine links between inter-model differences in ratios of ANO_x emissions, AVOC emissions, or 501 the transport tracers, with the ratio of PAN source-receptor relationships from two intercontinental regions. 502 At Mount Bachelor, the EU and EA source regions contribute similar amounts to multi-model mean PAN 503 during April (Figure 5). Across the individual models, however, the ratio of the EU to EA source region on PAN at Mount Bachelor varies from less than half to a factor of two (Figure 8, left column). We find that 504 505 the ratio of PAN attributed to the EU versus EA source regions at Mount Bachelor correlates strongly across the models with the ratio of the anthropogenic volatile organic compound (AVOC) emissions in the 506 respective source regions (r = 0.8; Figure 8). In contrast, the ratio of EU:EA anthropogenic emission 507 influence on PAN at Mount Bachelor shows little correlation with the respective regional NO_x emissions 508 509 used in the models, or with the differences in the simulated transport tracers (r=0.3 for both cases; Figure 8). As at Mount Bachelor, the model spread in the contribution to total simulated PAN from the EA versus 510 NA source regions at Jungfraujoch depends most on the regional AVOC ratios (r = 0.8; Figure 8 right 511 512 column), with little correlation with inter-model differences in NA:EA ANO_x emissions. Some correlation appears to exist between the NA:EA source-receptor relationships for PAN and the NA:EA transport 513 514 tracers (r = 0.6; Figure 8), although this weakens (to r = 0.5) if we exclude the left-most outlier. Finally, the individual symbols in Figures 7 and 8 depict the meteorological fields used in each model; there is no 515 516 obvious relationship between PAN source-receptor relationships and the choice of meteorological fields, as 517 expected from our analysis of the idealized CO tracers.





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519 For source-receptor pairs with significant relationships in Figures 6-8, the inter-model differences generally 520 correlate most strongly with differences in model AVOC emission inventories. What we attribute here to AVOC emissions actually represents the combination of those emissions plus the subsequent chemistry that 521 those emissions undergo prior to producing PAN. Even if models use the same AVOC emissions, 522 523 differences in chemical mechanisms are likely to contribute to inter-model differences in PAN (Emmerson 524 and Evans, 2009). The AVOC emissions implemented in the HTAP1 models range by a factor of 5 for NA and EA, and nearly a factor of 10 for EU. Our results indicate that constraints on the amount of VOC 525 526 emitted, and the PAN produced during oxidation of these VOCs, would narrow the range across model 527 estimates for PAN source-receptor relationships during April. Fischer et al. (2014) also emphasize the 528 sensitivity of global PAN budgets to AVOC emission inventories, which do not frequently receive much attention in terms of model evaluation. Independent methods to derive source-receptor relationships from 529 530 observations, such as the trajectory-based approach in Section 4 are also needed to evaluate model 531 responses to emission changes, as opposed to abundance-based evaluations. Finally, while model 532 differences in transport may not contribute most to the multi-model spread in the monthly mean source-533 receptor relationships examined here, we expect that transport would play a role in inter-model differences 534 in resolving individual trans-oceanic transport events.

535 6. Links between PAN and O₃ source-receptor relationships

We address here the extent to which observational constraints on PAN source-receptor relationships might also serve to narrow the range of uncertainty in the inter-model spread in intercontinental source-receptor relationships for O_3 (e.g., Fiore et al., 2009). We expect some commonality between the sensitivity of PAN and O_3 to changes in precursor emissions because (1) both species are produced from chemical reactions involving NO_x and VOC, and (2) PAN serves as a NO_x reservoir, which upon decomposition releases NO_x that can then produce O_3 far downwind of the region where the PAN (and O_3) precursors were originally emitted.

543

544 Earlier work found that inter-model differences in intercontinental influences on continental-scale surface O₃ in the HTAP1 models correlate with the wide range of AVOC emissions used in the models, particularly 545 546 over the EU (Fiore et al., 2009). Our analysis in Section 5 also showed a dependence of inter-model variations in the amount of PAN produced by (and transported from) a region on the magnitude of the 547 548 regional AVOC emissions. When we repeat the analysis in Figure 6 but for O₃, considering the influence 549 of the three source regions on two mountain sites (six total source-receptor pairs), we find little relationship between the spread in the simulated O_3 source-receptor relationships and in the magnitude of the AVOC 550 551 emissions from the source region. Similar to our findings for PAN, the O₃ source-receptor relationships 552 show little correlation with the model range in regional ANO_x emissions. Model differences in transport, as 553 diagnosed with the regional idealized CO tracers, are generally not correlated with the inter-model spread





in PAN or O_3 source-receptor relationships. The correlation with these tracers, although weak, is higher for O₃ than for PAN for all source-receptor pairs. A stronger correlation with the CO tracers does occur for the influence of EU emissions on O₃ at Jungfraujoch. Overall, this analysis supports our hypothesis that PAN is more sensitive to changes in emissions (and subsequent chemistry), particularly for VOC precursors, than O₃.

559

560 We now assess more directly the extent to which the inter-model range in source region influence on mountaintop PAN levels (Figure 3) is relevant for interpreting O_3 source-receptor relationships in the 561 562 HTAP1 models. We maintain our focus on April, when the signal from anthropogenic precursor emissions on PAN is at its seasonal peak according to the models. Figure 9 illustrates our correlation analysis of PAN 563 and O₃ source-receptor relationships at Jungfraujoch and Mount Bachelor across the 12 models (SR1-564 SR6XX where XX = EA, EU, or NA). Relationships vary across the individual source-receptor pairs, with 565 the inter-model variability in PAN explaining 16-60% of the inter-model differences in O₃ at the mountain 566 sites. The strongest relationships occur for the influence of regional sources on these mountain sites: NA at 567 568 Mount Bachelor and EU at Jungfraujoch. Intercontinental source-receptor pairs for O₃ and PAN at Mount Bachelor are also significant to within 90%, with variability in the PAN attributed to intercontinental 569 source regions explaining 25-35% of the variability in the corresponding O₃ source-receptor relationship. 570

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The correlations between PAN and O₃ source-receptor relationships could reflect a role for PAN transport 572 573 in contributing to O₃ production over the receptor region, or may instead reflect co-production of PAN and 574 O₃ from oxidation of regional precursor emissions followed by transport in the same air mass. In the latter case, PAN is serving as a proxy for O3 transport whereas in the former case, PAN is serving as the actual 575 pathway by which O₃ is transported. We do not have model diagnostics that allow us to distinguish 576 between these two roles for PAN. The correlations between PAN and O₃ source-receptor relationships, 577 578 however, suggest that long-term PAN measurements contain signals relevant for constraining the relative 579 importance of regional vs. intercontinental emissions on both PAN and O₃.

580 7. Conclusions and recommendations

581 We investigated monthly mean PAN mixing ratios from a suite of multi-model sensitivity simulations 582 coordinated by the Task Force on Hemispheric Transport of Air Pollution (HTAP, 2007; 2010; Table 1) in conjunction with mountaintop measurements (Table 3) to quantify the influence of regional versus 583 584 intercontinental anthropogenic emissions on PAN at these sites. PAN transport and subsequent 585 decomposition leads to O₃ production, one pathway for intercontinental O₃ transport. Correlations between 586 inter-model differences in PAN and O₃ responses to perturbations in regional anthropogenic emissions 587 (e.g., EU at Jungfraujoch and NA at Mount Bachelor; Figure 9) imply that constraints on PAN should 588 reduce uncertainty in estimates of O_3 source-receptor relationships. As the O_3 response to climate warming





is partially dependent on thermally-sensitive PAN (e.g., Doherty et al., 2013), better constraints on PAN are also needed to increase our confidence in the projected responses of O_3 to climate change.

591

592 The wide range in PAN levels, and its source-receptor relationships, across the HTAP1 models indicates a need to improve our understanding of the chemical and physical mechanisms underlying PAN distributions 593 594 and seasonal cycles in the troposphere. Uncertainties in emissions as well as yields of PAN from VOCs, 595 both natural and anthropogenic, hinder a full understanding of PAN in the troposphere (Figure 6; see also Fischer et al., 2014). Regional AVOC emissions usually differ more than NO_x emissions (e.g., Figures 7 596 597 and 8), and have fewer observational constraints (e.g., NO₂ columns are routinely retrieved from space). 598 The range in model representations of VOC chemistry also contributes to inter-model differences in PAN 599 (e.g., Emmerson and Evans, 2009).

600

601 To further our understanding of PAN in the troposphere, we recommend rigorous process-oriented evaluation of models, both event-based (e.g., Fischer et al., 2010; Alvarado et al., 2010) and long-term 602 603 statistical evaluations of PAN and its relationships with other species. Ongoing multi-model projects, 604 including HTAP Phase 2 (HTAP2), the Chemistry-Climate Model Initiative (CCMI), and the upcoming 605 AerChemMIP simulations, may prove useful for some of this analysis. Accurate diagnosis of the impact of differences in NO_x and VOC emissions on PAN and O₃ source-receptor relationships, however, hinges on 606 607 the availability of dedicated source attribution diagnostics or emission perturbation simulations, and would 608 also benefit from the availability of chemical budget diagnostics.

609

610 Multi-year PAN records at existing measurement sites would provide much needed constraints on PAN 611 production and export from the three major anthropogenic source regions at northern mid-latitudes (East Asia, Europe, and North America). We identified only five multi-year datasets at mountain sites, four of 612 613 which are located near each other in Europe, and only one of which continues at present (Schauinsland). In 614 addition to the sites in Table 3 where PAN has been measured for at least two years, we suggest that PAN 615 measurements at other existing sites are needed to better constrain export from the North American and Asian continents and subsequent trans-oceanic transport. The Tropospheric Ozone Assessment Report 616 617 (TOAR) database (Schultz et al., 2017) contains O₃ measurements from more than 60 stations at altitudes 618 above 2500 m, a useful starting point for identifying sites where observational constraints would best discriminate across models (e.g., where the models vary most in Figure 1). Adding PAN to the suite of 619 620 species measured routinely aboard commercial aircraft (Ekstein et al., 2017; Petzold et al., 2015) would also enable knowledge of PAN distributions across much of the northern mid-latitude free troposphere. 621 Although in their infancy, new satellite products for PAN offer much potential for documenting PAN 622 623 distributions, particularly in the upper troposphere, as well as temporal variability and trends across the globe (e.g., Fadnavis et al., 2014; Jiang et al., 2016; Payne et al., 2014; 2017; Zhu et al., 2015; 2017). 624 625 Sampling models in a fully consistent manner with satellite-derived tropospheric PAN columns, along





aircraft flight tracks, or at long-term monitoring sites requires archiving model fields at high temporalfrequency at the measurement locations.

628

629 The substantial inter-model differences in PAN abundances and their attribution to regional anthropogenic 630 emissions identified here highlight the potential for observations of PAN to provide much-needed 631 constraints on model estimates of intercontinental transport. Seasonal and inter-annual variability in the 632 signal strength from anthropogenic (largest in spring at northern mid-latitudes) versus other sources 633 suggests process-level information for testing models can be obtained from careful interpretation of 634 seasonal cycles. Changes in meteorology and biomass burning (Fischer et al., 2011; Zhu et al., 2015) such 635 as those driven by ENSO (Koumoutsaris et al., 2008), as well as biogenic and lightning sources (Payne et 636 al., 2017), will also cause PAN to vary from one year to another and are expected to change as climate 637 warms. Springtime PAN measurements are most likely to resolve model discrepancies in anthropogenic 638 emission influences on PAN, whereas summertime measurements are strongly influenced by biogenic, 639 lightning, and biomass burning influences. We conclude that long-term observations of PAN in the free 640 troposphere during spring, although currently limited, could be a keystone measurement for improving 641 model simulations of intercontinental pollutant transport.

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COIFOMAA) used in t	nis study.					
Model	Resolution	Institute	Model	SR1	SR6xx	COfromXX
	(lat-lon-		contact			
	layers)					
CAMCHEM-	AMCHEM- 2.5°x2°x30 NCAR, USA		Peter Hess	Х	Х	Х
3311m13						
FRSGCUCI-v01	2.81°x2.81°	Lancaster	Oliver Wild	Х	Х	Х
	x37	University, UK				
GEMAQ-v1p0	2°x2°x28	York	Alex Lupu	Х	Х	Х
		University,				
		Canada				
GEOSChem-v07	2.5°x2°x30	Harvard	Rokjin Park	Х	Х	Х
		University,				
		USA				
GISS-PUCCINI-	5°x4°x23	NASA GISS,	Drew	Х	Х	Х
modelE		USA	Shindell			
GMI-v02f	2.5°x2°x42	NASA GSFC,	Bryan	Х	Х	Х
		USA	Duncan			
LMDZ3-INCA1	3.75° x 2° x	CEA, France	Sophie	Х	Х	
	19		Szopa			
LLNL-IMPACT-	2.5° x 2° x	LLNL, USA	Dan	Х		
T5a	48		Bergmann			
MOZARTGFDL-	1.88° x	NOAA GFDL,	Arlene Fiore	Х	Х	Х
v2	1.88° x 28	USA				
MOZECH-v16	1.88° x	FZ Julich,	Martin	Х	Х	Х
	1.88° x 28	Germany	Schultz			
STOC-HadAM3-	5° x 5° x 19	University of	Ruth	Х	Х	Х
v01		Edinburg, UK	Doherty,			
			David			
			Stevenson			
STOCHEM-v02	3.75 x 2.5°	Met Office,	Bill Collins,	Х		
	x 20	Hadley Center,	Michael			
		UK	Sanderson			
TM5-JRC-cy2-	1° x 1° x 25	JRC, Italy	Frank	X	X	X
ipcc-v1			Dentener			

902 Table 1: Specifications of individual models contributing to the HTAP1 simulations (SR1, SR6xx, and COfromXX) used in this study.





	UM-CAM-v01	3.75° x 2.5°	University of	Guang Zeng	Х	Х	Х
		x 19	Cambridge, UK				
904					1		
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917 Table 2: Simulations from the HTAP1 multi-model intercomparison used in this study.

Simulation	Description				
SR1	Base case (see Section 2.1 for details)				
SR6EA	SR1 but with anthropogenic emissions of all O ₃ precursors (NO _x +CO+NMVOC)				
	and aerosols within EA decreased by 20%				
SR6EU	SR1 but with 20% emissions reductions within the EU region				
SR6NA	SR1 but with 20% emissions reductions within the NA region.				
ISOPNA	SR1 but with a 20% increase in NA isoprene emissions (MOZARTGFDL-v2 model				
	only)				
COfromEA	Idealized tracer simulation in which all models use identical CO emissions, emitted				
	within the EA region, with a 50-day e-folding lifetime.				
COfromEU	Same as COfromEA but for the EU region.				
COfromNA	Same as COfromEA but for the NA region.				

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923 Table 3: PAN observations at mountain sites in the northern hemisphere mid-latitudes used in this study.

Site	Location	Elevation	Measurement Period (s)	Reference (s)			
Mount Bachelor	43.979° N, 121.687° W	2763m	3 April – 18 June 2008, 30 August – 7 October 2008, 26 March – 20 May 2009, 23 March – 25 May 2010	(Fischer et al., 2010;Fischer et al., 2011)			
Hohenpeissenberg	47.80° N, 11.02° E	985 m	January 2003 – December 2008	http://www.dwd.de/de/GAW (Gilge et al., 2010)			
Jungfraujoch	46.55°N, 7.98°E	3580 m	April 1997 – May 1998, Aug 30 2005 – Sept 16 2005, Throughout 2005, but not continuous	(Balzani Lööv et al., 2008;Carpenter et al., 2000;Zellweger et al., 2000;Zellweger et al., 2003)			
Zugspitze	47.42° N, 10.98° E	2960 m	May 2004 – December 2008	http://gaw.kishou.go.jp			
Schauinsland	47.92° N, 7.92° E	1205m	January 1995 – December 2010	www.umweltbundesamt.de			

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930 Figure 1: Multi-model ensemble (n=14; Table 1) average PAN mixing ratios (ppt; left panels) and relative 931 standard deviation (the absolute standard deviation across the models divided by the ensemble mean; right 932 panels) at 650 hPa in April and July; relative standard deviations are masked out (white) for regions where 933 multi-model mean PAN falls below 100 ppt. The models were sampled at 650 hPa by vertically interpolating 934 between the bounding grid cells and then re-gridded horizontally to a common 1°x1° grid. White lines denote 935 the HTAP1 source regions: Europe and North Africa (EU), East Asia (EA) and North America (NA) from left to 936 right, and white circles indicate the five mountain sites used in our analysis (Zugspitze and Hohenpeissenberg 937 are too close to differentiate on the map; see Table 3).

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Figure 2: Monthly mean observed (black symbols for the multiyear mean, with minimum and maximum across the available years shown as black vertical lines; see Table 3) and simulated (the 14-model mean is shown as the grey dashed line; grey shading encompasses maximum and minimum values from any model in each month) PAN at northern mid-latitude mountain sites with multiple years of PAN measurements. For illustrative purposes, we show a "high" model (CAM-Chem, brown), a "low" model (GISS-ModelE, green), and two models that vary from site to site in their comparison to the other models (MOZARTGFDL, red and GMI, cyan). Also shown are year 2001 monthly mean values available at the Schauinsland site (blue triangles offset by a few days for clarity) and the standard deviation of measurements within each month (blue vertical lines).







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954 Figure 3: Absolute (left) and fractional (right) mixing ratios of PAN at Jungfraujoch attributed to EU (top), NA 955 (middle) and EA (bottom) precursor emissions by the HTAP1 models ((SR1-SR6XX)*5 for comparison with 956 trajectory-based estimates; see Table 1). Shown are the 12-model mean (dashed), median (dot-dashed) and range (gray shading). Selected individual models are colored as in Figure 2. Also shown are the trajectory-957 958 based attributions (solid triangles; Section 2.3). See Section 4 for details.

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964 Figure 4: Contribution to PAN mixing ratios at Jungfraujoch from EU (black), NA (red) and EA (blue) 965 anthropogenic precursor emissions and from NA isoprene emissions (green) in the MOZARTGFDL model. We 966 take the difference between the base simulation and one in which emissions are decreased by 20% for the 967 anthropogenic cases but increased by 20% for isoprene, and then multiply each difference by 5 to estimate a 968 100% contribution.

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Figure 5: Multi-model monthly mean PAN mixing ratios (black circles) at Mt. Bachelor (left) and Jungfraujoch
(right). We take the difference between the base simulation and one in which emissions are decreased by 20%
and then multiply the difference by 5 to estimate a 100% contribution associated with anthropogenic precursor
emissions from Europe (green), North America (red), East Asia (blue); the sum of the anthropogenic
contribution from these three regions is also shown (dashed black) for comparison with total simulated PAN
(solid black).

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Figure 6: Source-receptor relationships for PAN (ppt) at Jungfraujoch (top) and Mt. Bachelor (bottom) in each
 HTAP1 model (diagnosed as the difference between the SR1 and SR6xx simulations in Table 1) versus the
 annual emission of anthropogenic VOC (AVOC; Tg C a⁻¹) within the NA (left), EU (middle) and EA (right)
 source regions. The Spearman rank correlation coefficient (more robust to outliers than the traditional Pearson
 coefficient) and associated p-value are shown in each panel.

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998 Figure 7: The ratio in April of simulated source-receptor relationships for PAN sampled at Mt. Bachelor (left) 999 or Junfraujoch (right) produced from emissions within the continental region where the site is located versus an 1000 intercontinental region, plotted as a function of the regional-to-intercontinental ratio of anthropogenic NO_x 1001 emissions (ANOx; top row), of anthropogenic volatile organic compound emissions (AVOC; middle row), or of 1002 idealized tracers of model transport (COfromXX where XX = region of origin). Each point corresponds to an 1003 individual model and is color-coded by the meteorological fields used in the simulation: blue triangles for GEOS 1004 winds; red circles for NCEP; black diamonds for ECMWF; cyan upside-down triangles for CMC; green squares 1005 for general circulation models forced by observed sea surface temperatures and sea ice. At Jungfraujoch, one 1006 model with a ratio of 8.3 for EU to NA PAN is excluded. The Spearman rank correlation coefficient is shown in 1007 each panel.















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1020 Figure 9: Source-receptor pairs for O3 versus PAN at Jungfraujoch (top) and Mt. Bachelor (bottom), obtained 1021 from the difference between the SR6XX and SR1 simulations available from 12 models (Table 1), where XX 1022 denotes the NA (left), EU (middle) or EA (right) source region. Each model thus contributes one point in each 1023 panel. Also shown are the Spearman (rank) correlation coefficient and p-values.